

Robustness of Model-Based Risk Reduction Strategies

Julian Richardson

RIACS, NASA Ames Research Center

julianr@riacs.edu

Daniel Port

University of Hawaii

dport@hawaii.edu

Martin Feather

Jet Propulsion Laboratory, California Institute of Technology

Martin.S.Feather@jpl.nasa.gov

This research was carried out at the NASA Ames Research Center, the University of Hawaii, and the Jet Propulsion Laboratory, California Institute of Technology, partly under a contract with the National Aeronautics and Space Administration and partly supported by JAXA

Introduction

- Models can predict effect of future actions; used to make decisions:
 - more reliably than human judgement,
 - more flexibly than fixed processes
 - e.g. use of COCOMO II or other software cost models is mandatory at NASA (NPR 7150.2)
- Risk reduction (including V&V) is critical and expensive part of projects
- Improvements in risk reduction can save money and/or reduce risk
- Model-based choice of risk reduction strategies:
 1. *Quantify risk* in each risk category
 2. *Quantify cost and risk reduction* for each technique
 3. Choose optimal combination of risk reduction techniques
- *Q: How much does effectiveness of chosen strategy depend on accurate quantification of risks & mitigations?*
 - ran sensitivity analysis experiments for different optimization strategies
- *A: Not much*
 - *optimized strategy nearly always beats any fixed strategy*

Optimizing Risk Reduction

- We consider here two different algorithms
- *Strategic Method* (Port, Kazman et al)
 - employed with JAXA case studies
 - algorithm gives provably optimal risk reduction strategies
 - as long as assumptions hold
 - well suited to independent V&V (IV&V)
- *Defect Detection and Prevention* (Cornford & Feather)
 - design-level identification and mitigation of system/software risks
 - developed at JPL, used for many NASA mission technologies
 - rapid elicitation of relationships between objectives, risks, mitigations
 - risks harm objectives,
 - mitigations reduce risks
 - uses a standard of heuristic search (simulated annealing) to make near-optimal selections from among dozens – hundreds of mitigations

Strategic Method

- Inputs:

- loss potential and probability for each attribute (risk)

Attribute (i)	A1	A2	A3	A4
Loss potential for Ai	100	90	90	80
$P_{\text{before}}(Ai)$	6	5	20	15

- cost and reduction in loss probability applying each technique to each attribute

Cost / resultant loss probability Assessing Ai with Tj	A1	A2	A3	A4
T1	50 / 4	NA	10 / 15	70 / 12
T2	100 / 6	NA	NA	100 / 13
T3	NA	NA	80 / 15	80 / 12

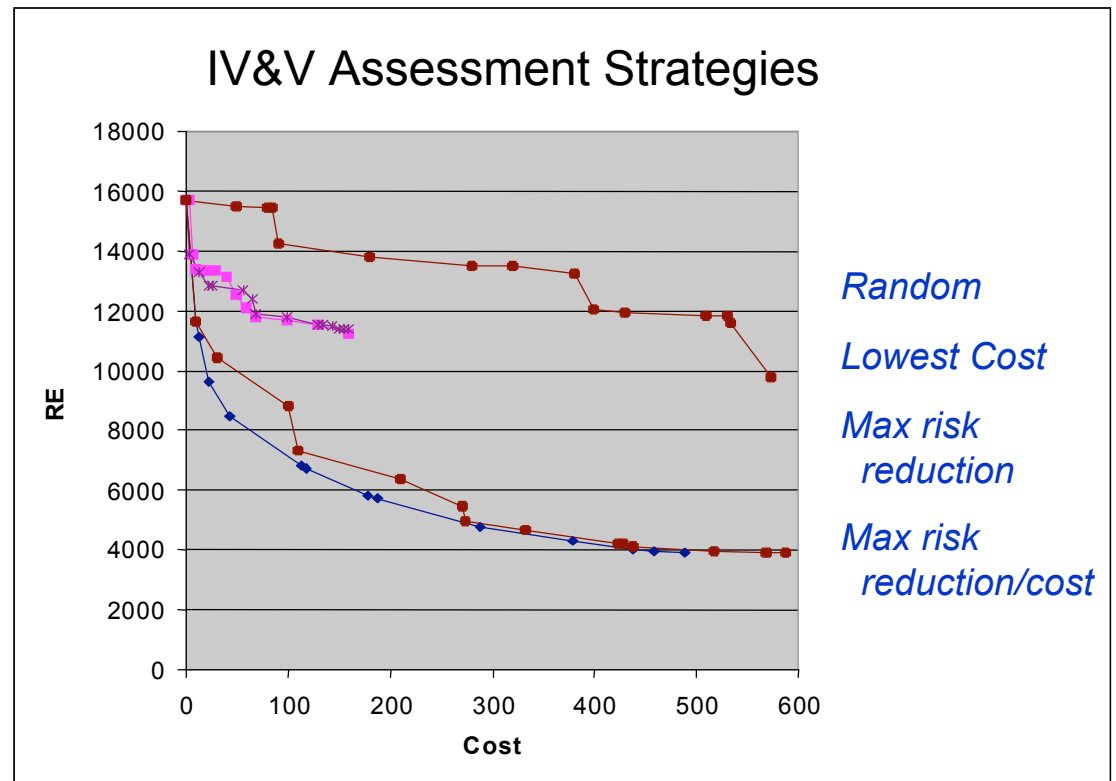
- Output:

- optimal order to apply techniques to attributes for any budget

Attribute	Technique	RE Change	Cost	CB	risk reduction	cumulative cost
None	None	None	None	None	15700	0
A13	T11	4050	10	405	11650	10
A7	T9	500	3	166.667	11150	13
A11	T11	1500	10	150	9650	23

Risk Reduction vs Cost

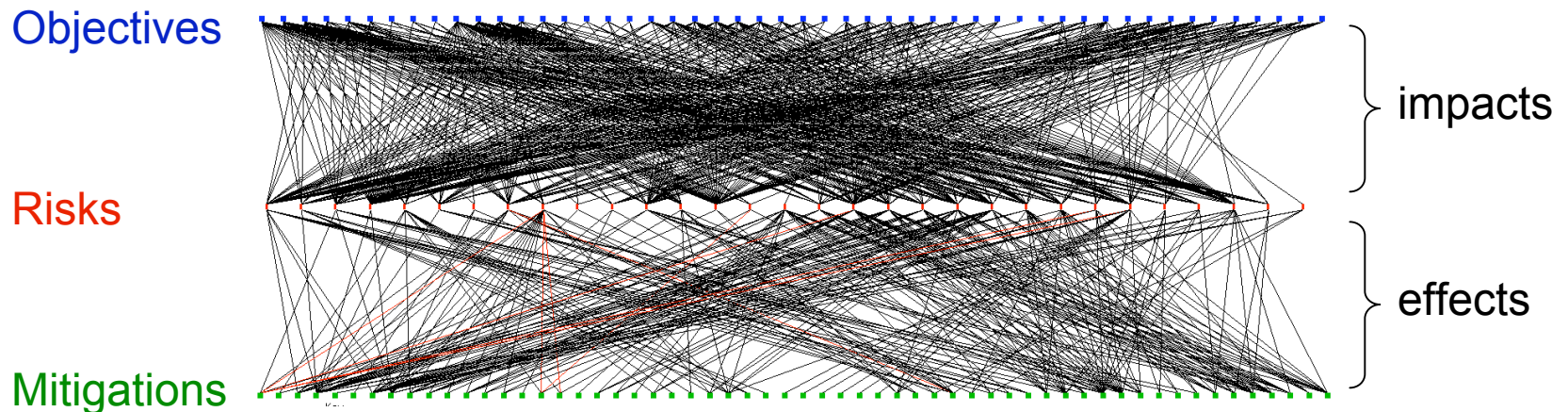
- Graph on right plots risk exposure vs cost for various strategies.
- The benefit provided by strategy is its risk reduction.
- Better strategies produce more benefit for given cost → have lower curves.



DDP

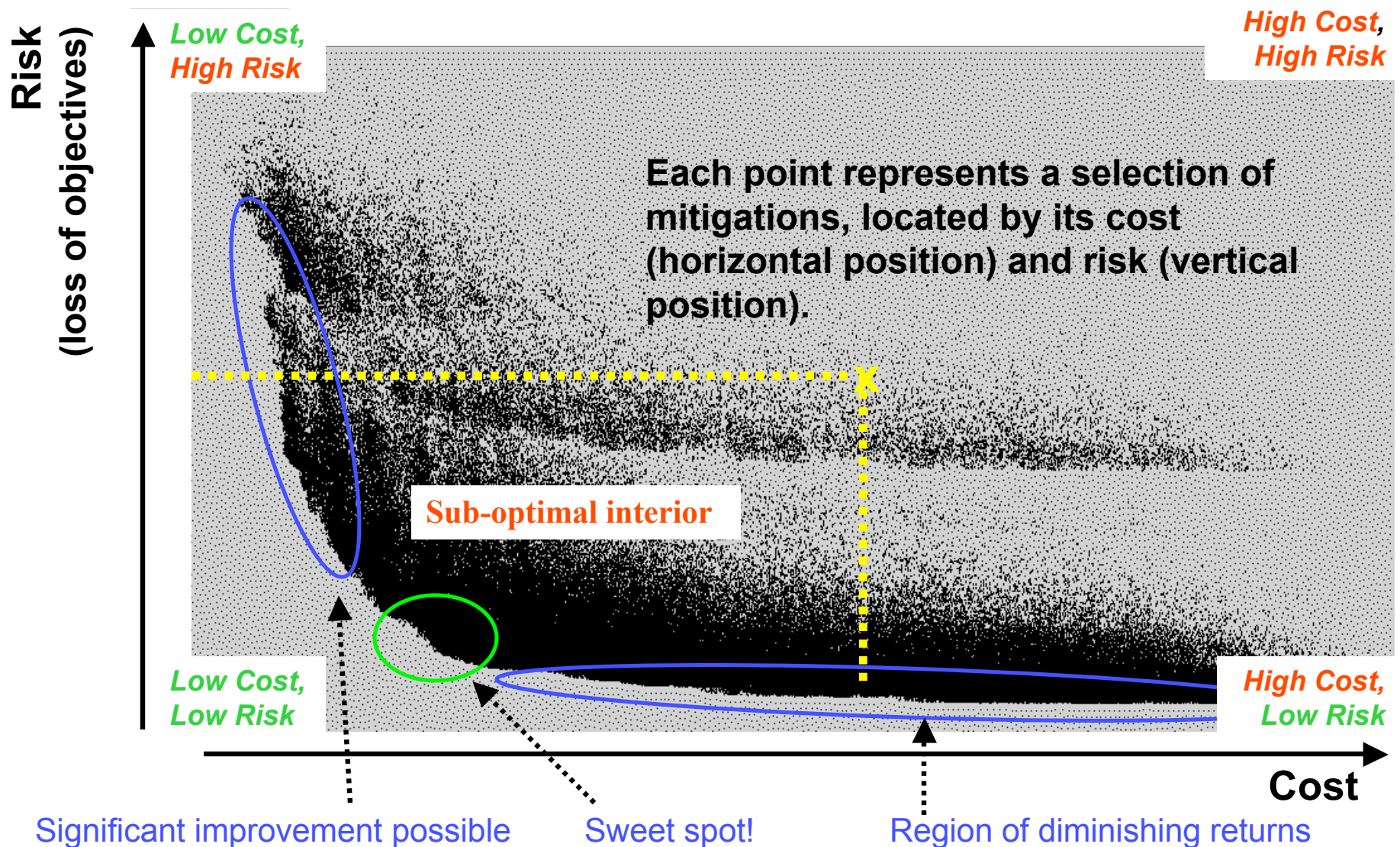
- Inputs:
 - amount ($0 \leq \text{impact} \leq 1$) by which each risk reduces each objective
 - amount ($0 \leq \text{effect} \leq 1$) by which each mitigation reduces each risk
 - cost of each mitigation
 - total budget
 - Output:
 - heuristically optimized (maximal attainment of objectives) selection of mitigations for that budget
-

This is the topology of the connections for an actual application of DDP
– note that associated with each line is the amount (impact or effect):



DDP method results

58 mitigations = 2^{58} (approx 10^{17}) ways of selecting: searches using “simulated annealing”, extended across entire cost range to reveal cost/risk tradespace



Need for Sensitivity Analysis

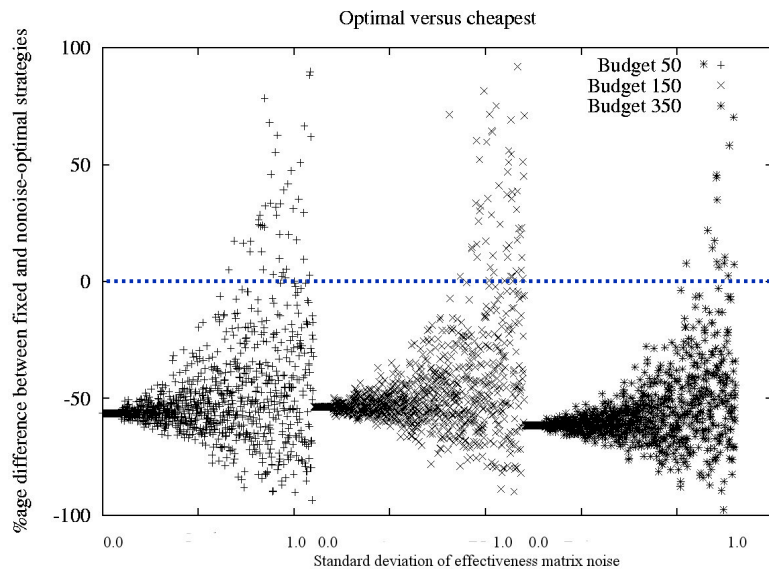
- Algorithms optimize strategy selection given knowledge of magnitudes and costs of risks and mitigations.
 - Hard to know these things in advance –
does (in)accuracy effect validity of decisions reached by applying these algorithms?
 - Questions:
 1. *How much does effectiveness of chosen strategy depend on accurate quantification of risks & mitigations?*
 - experiments to vary actual from specified effectiveness or risk
 2. *How much does optimized strategy improve on fixed* strategy?*
 - experiments evaluate difference between optimized strategy, and each of four kinds of fixed strategy (a) random, (b) “reasonable”, (c) cheapest, (d) “great” (optimal for nominal risk levels)
- * Fixed strategy = for a given budget, a predetermined selection of mitigations that is the same no matter the problem

Sensitivity wrt Effectiveness Matrix

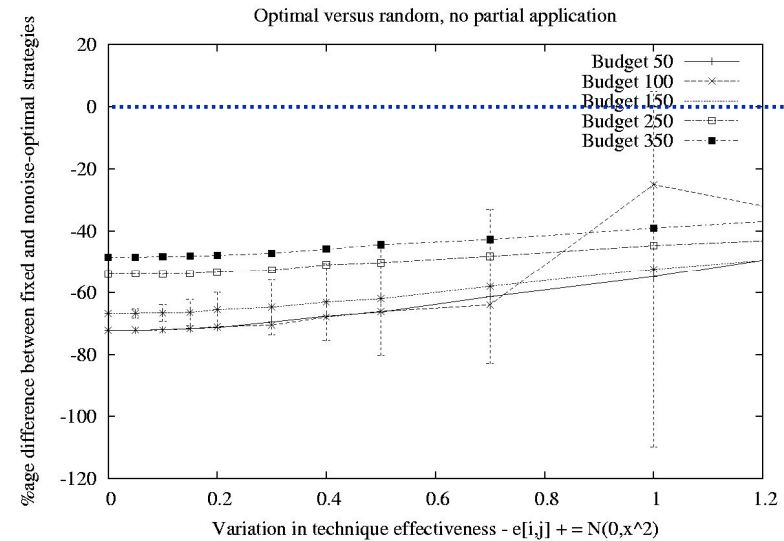
- *Knowledge of risk or technique effectiveness is often uncertain.*
 - Compute effectiveness matrix: effectiveness of technique T_i on attribute A_j , $\rho_{ij}^0 = (P_{before}(A_j) - P_{after}(A_j, T_i)) / P_{before}(A_j)$
1. Repeatedly:
 - a) Pick fixed budget $b \in \{50, 100, 150, 200, 250, 300, 350, 400\}$
 - b) Pick noise level $\sigma \in \{0, 0.05, 0.1, 0.15, 0.2, 0.3, 0.4, 0.5, 0.7, 1.0\}$
 - c) $N (=1000)$ times:
 - i. Add noise to effectiveness matrix: $\forall i, j, \rho_{ij} = \|\rho_{ij}^0 + N(0, \sigma^2)\|$
 - ii. Evaluate %age difference between S_{opt} and S_{fixed} , $\Delta = (\delta RE(b, S_{fixed}) - \delta RE(b, S_{opt})) / \delta RE(b, S_{opt})$
 - d) Add a point to the plot with x coordinate σ , y coordinate the mean value of Δ , and if desired add error bars to that point to indicate the standard deviation in Δ

Results

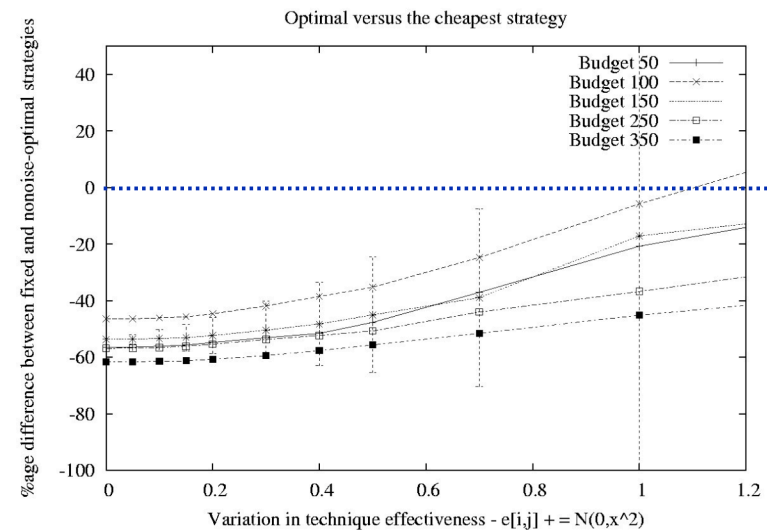
- Optimized strategy:
 - significantly better than random strategy
 - significantly better than cheapest strategy
 - even for inaccurate knowledge of effectiveness or risk



Optimized is better
Optimized is worse



Optimized is better
Optimized is worse



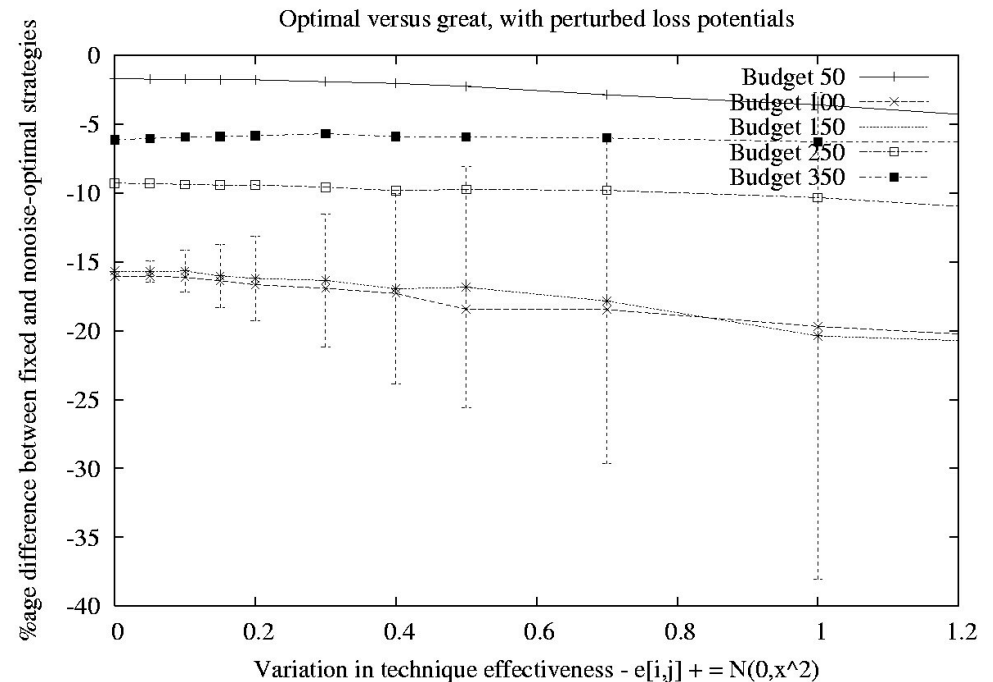
Optimized is better
Optimized is worse

Comparison with Optimal Fixed Strategy

- *Strategic Method adapts strategy to risk profile*
 - *maybe we could just optimize once, for typical risk profile?*
1. Use Strategic Method to calculate optimal strategy S_{fixed} given loss potentials, L_j for each attribute A_j
 2. For each L_j randomly choose L'_j from $\{0, L_j, L_j \times 1.5\}$
 3. Use Strategic Method to calculate optimal strategy S_{opt} for perturbed loss potentials
 4. Perform sensitivity analysis wrt effectiveness matrix as before

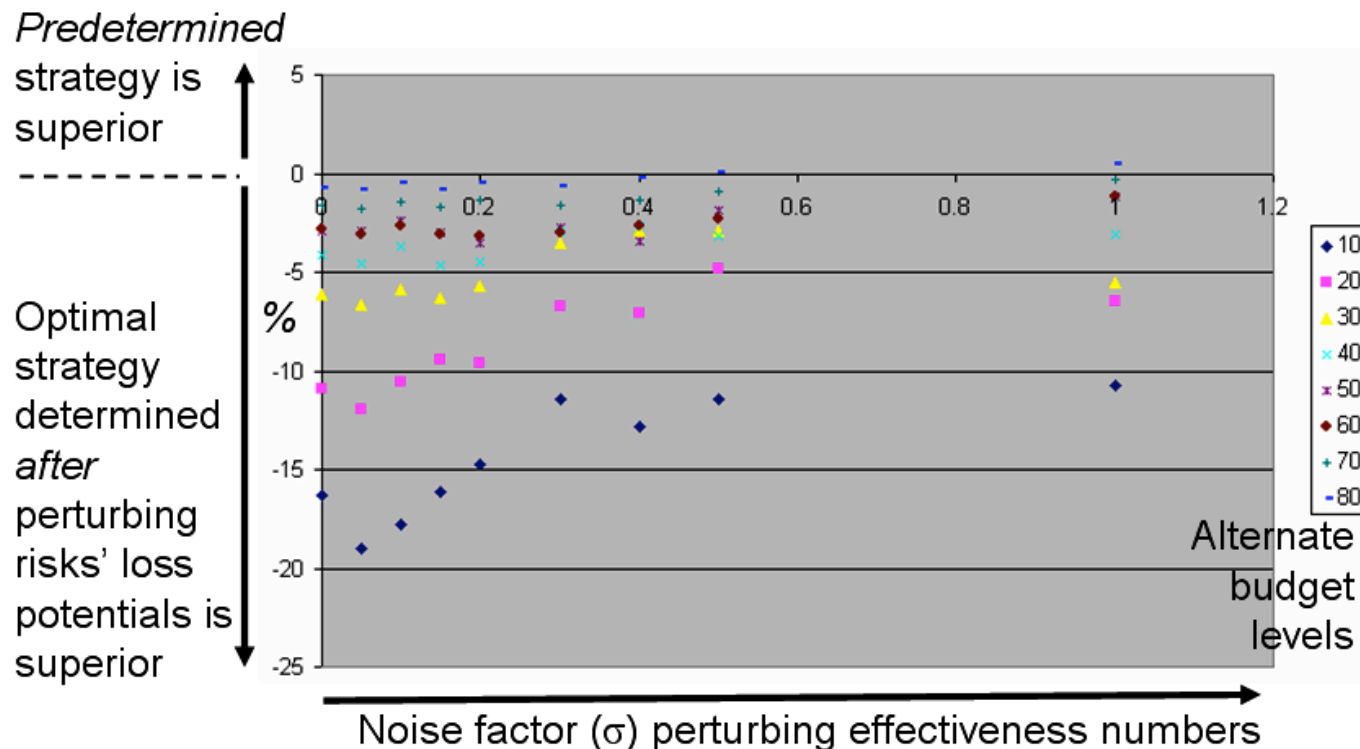
Results

- Graph shows result from a typical choice of perturbed loss potentials.
- Optimal significantly better than 'great'
- *Fixing a strategy to that which is optimal for a typical risk profile is usually **inferior to an optimization based on the estimated risk profile**, even in the face of inaccuracies in those estimates.*



Sensitivity analysis of DDP

- DDP uses different optimization algorithm (simulated annealing) and different calculation of risk reduction.
- Strategic method experiments repeated with DDP* - **conclusions for Strategic Method hold for DDP too:**



*DDP changed to allow fractional application of the final V&V strategy, as per Strategic Method

Conclusions

- Sensitivity analysis of strategic method and DDP wrt knowledge of technique effectiveness and risk reduction
- *Optimized strategy is better than alternatives even when significant uncertainty exists in estimates of effectiveness and risks.*
- *Significant cost reductions or risk reductions are achievable:*
 1. *Estimate magnitude of risks + effectiveness & costs of available mitigations.*
 2. *Choose optimal strategy (e.g. using Strategic Method).*